

eBird API Tutorial

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1 Tutorial: Using the eBird API for Digital Conservation

This tutorial teaches how to access eBird observations data through the citizen science platform’s Application Programming Interface (API) and illustrates a very basic use of such data for conservation analysis.

The tutorial is designed for a wide range of users. For those with minimal training in programming, the entire notebook can be “played” without writing a single line of code. For those with experience in the Python programming language, the notebook can be copied and adjusted. Those unfamiliar with “digital conservation” will learn a bit more about an important tool by which species can be tracked, while those already in the conservation sector will have the chance to engage with this familiar dataset in a new way.

1.1 eBird

First, we introduce eBird. eBird is a citizen science platform for bird species tracking and identification. Users report sightings of birds and upload them to the platform, which displays these observations on a map anyone can access. From a user perspective, this provides a digital way birders can manage their checklists. From a researcher and conservationist perspective, eBird provides observational data in volumes that no single team could ever possibly hope to acquire. eBird data can help illustrate migration patterns, where rare and threatened species are found, and show us where people are interacting with nature.

1.1.1 eBird’s API

An API is a standardized, structured set of names and codes that enable the public to access platform data. Most platforms provide an API, though they differ in how much access they provide. For instance, X (formerly Twitter), has an API that allows researchers to search for tweets based on keywords, geographic location, hashtags, and more and to download these without ever actually opening the website and using its search tool.

Each platform’s API is unique and what can be accessed how varies. The only way to know is to review the API documentation. eBird’s is available [here](#). These are often not easy to read unless you have a background in programming as well as extensive familiarity with the platform and its data.

For our purposes, we are interested in observations, so we open the `data/obs` folder and select **Recent observations in a region**. This provides us with the general template we need to follow to access relevant eBird data. The query parameters are variables we can adjust in order to access the exact kind of data we want.

In the following cell of code, we import some code that will help us later (`requests`) and then “call up” the eBird API. Basically what we are doing is using programming to “visit” a website, one that contains the data we want, but in the JSON format. To confirm this is what’s happening, try entering the url into a browser.

Note: many APIs require you to create an account first and/or get a “token” (password). This may be separate from your platform account. To access eBird’s data through its API, you will need to get an API token. You can do so [here](#).

```
[4]: import requests

url = "https://api.ebird.org/v2/data/obs/KZ/recent"

payload={}
headers = {
    'X-eBirdApiToken': '{{x-ebirdapitoken}}' # Replace {{x-ebirdapitoken}} with
    ↳ your API token
}

response = requests.request("GET", url, headers=headers, data=payload)

print(response.json()[0:5])
```

```
[{'speciesCode': 'brambl', 'comName': 'Brambling', 'sciName': 'Fringilla
montifringilla', 'locId': 'L16827294', 'locName': '  --
(
    ) [Atreyu--TCO Village area (Ural River banks)]',
'obsDt': '2024-01-26 13:59', 'howMany': 9, 'lat': 47.099957, 'lng': 51.9071,
'obsValid': True, 'obsReviewed': False, 'locationPrivate': False, 'subId':
'S159803932'}, {'speciesCode': 'hoocro1', 'comName': 'Hooded Crow', 'sciName':
'Corvus cornix', 'locId': 'L16827294', 'locName': '  --
(
    ) [Atreyu--TCO Village area (Ural River banks)]',
'obsDt': '2024-01-26 13:59', 'howMany': 18, 'lat': 47.099957, 'lng': 51.9071,
'obsValid': True, 'obsReviewed': False, 'locationPrivate': False, 'subId':
'S159803932'}, {'speciesCode': 'gretit1', 'comName': 'Great Tit', 'sciName':
'Parus major', 'locId': 'L16827294', 'locName': '  --
(
    ) [Atreyu--TCO Village area (Ural River banks)]',
'obsDt': '2024-01-26 13:59', 'howMany': 3, 'lat': 47.099957, 'lng': 51.9071,
'obsValid': True, 'obsReviewed': False, 'locationPrivate': False, 'subId':
'S159803932'}, {'speciesCode': 'fieldf', 'comName': 'Fieldfare', 'sciName':
'Turdus pilaris', 'locId': 'L16827294', 'locName': '  --
(
    ) [Atreyu--TCO Village area (Ural River banks)]',
'obsDt': '2024-01-26 13:59', 'howMany': 1, 'lat': 47.099957, 'lng': 51.9071,
'obsValid': True, 'obsReviewed': False, 'locationPrivate': False, 'subId':
'S159803932'}, {'speciesCode': 'eurgre1', 'comName': 'European Greenfinch',
'sciName': 'Chloris chloris', 'locId': 'L16827294', 'locName': '  --
(
    ) [Atreyu--TCO Village area (Ural River
banks)]', 'obsDt': '2024-01-26 13:59', 'howMany': 1, 'lat': 47.099957, 'lng':
51.9071, 'obsValid': True, 'obsReviewed': False, 'locationPrivate': False,
```

```
'subId': 'S159803932']]
```

The above cell of code returns a list of recent observations in “KZ” (Kazakhstan). You might have to scroll a long ways to see it all! That’s because of how the eBird template code works. We can make some adjustments to it.

Below, we change our region to Ontario, Canada (CA-ON). We also format it better using the pandas programming package.

```
[ ]: import json
import pandas

# Here, we create a function that we can reuse later to get different kinds of
↳ data from eBird
def get_ebird_data(url):

    payload={}
    headers = {
        'X-eBirdApiToken': '{{x-ebirdapitoken}}' # Replace {{x-ebirdapitoken}} with
↳ your API token
    }

    response = requests.request("GET", url, headers=headers, data=payload) #
↳ "Visit" the website with this data
    #try:
    # data = json.loads(response.text) # Load the data in the JSON format
    #except:
    data = pandas.read_json(response.text)

    return data

data = get_ebird_data("https://api.ebird.org/v2/data/obs/CA-ON/recent") #
↳ Change region to CA-ON
data
```

```
[ ]:      speciesCode      comName      sciName      locId \
0      rebwoo      Red-bellied Woodpecker      Melanerpes carolinus      L385611
1      merlin      Merlin      Falco columbarius      L7944575
2      whbnut      White-breasted Nuthatch      Sitta carolinensis      L26394771
3      bkcchi      Black-capped Chickadee      Poecile atricapillus      L26394771
4      amegfi      American Goldfinch      Spinus tristis      L4140370
..      ...      ...      ...      ...
192     sprgro      Spruce Grouse      Canachites canadensis      L27290314
193     baisan      Baird's Sandpiper      Calidris bairdii      L418231
194     greegr      Great Egret      Ardea alba      L7
195     leasan      Least Sandpiper      Calidris minutilla      L382375
196     osprey      Osprey      Pandion haliaetus      L2175806
```

		locName	obsDt	\
0		Wildwood Reservoir	2023-12-05 11:55	
1	57-299 Ravenscrest Rd, Georgina CA-ON (44.2144,...		2023-12-05 11:54	
2		Heyden, Ontario	2023-12-05 11:52	
3		Heyden, Ontario	2023-12-05 11:52	
4	Queensville - Eves-Kinsley Residence		2023-12-05 11:50	
..		
192	Snake Falls Rd, Kenora, Unorganized CA-ON		2023-11-24 15:41	
193		Blenheim Sewage Lagoons	2023-11-24 11:24	
194		Holiday Beach CA	2023-11-24 09:00	
195	Harrow Sewage Lagoons (restricted access)		2023-11-23 16:00	
196	Burlington--Royal Botanical Gardens (Cherry Hi...		2023-11-23 09:14	

	howMany	lat	lng	obsValid	obsReviewed	locationPrivate	\
0	1.0	43.242202	-81.044083	True	False	False	
1	1.0	44.214377	-79.400671	True	False	True	
2	1.0	46.642828	-84.306234	True	False	True	
3	NaN	46.642828	-84.306234	True	False	True	
4	5.0	44.136378	-79.449421	True	False	True	
..	
192	1.0	50.832656	-93.463080	True	False	True	
193	1.0	42.319589	-82.020750	True	True	False	
194	1.0	42.034026	-83.040726	True	True	False	
195	1.0	42.044640	-82.940168	True	True	False	
196	1.0	43.294512	-79.878824	True	True	False	

	subId	exoticCategory
0	S155793499	NaN
1	S155793460	NaN
2	S155793474	NaN
3	S155793474	NaN
4	S155793575	NaN
..
192	S155126614	NaN
193	S155105425	NaN
194	S155130388	NaN
195	S155101251	NaN
196	S155064199	NaN

[197 rows x 14 columns]

1.1.2 API Documentation

At this point, it's worth reviewing the eBird API in a bit more detail to learn more about what exactly we're getting when we make this "request" to the platform API.

The API documentation states that "Results include only the most recent observation for each species in the region specified." So there won't be multiple osprey observations, as an example,

unfortunately. If we wanted to focus on a specific species and ALL recent observations of it, we could, but that would require using a different part of the API. Instead, this allows us to get a general sense of the range of species being observed in the region recently.

Another parameter is `back`, which specifies how far back the API will go into the eBird database to retrieve results. The default is 14 days but we can set it to up to 30 by changing this in our list of headers.

```
[ ]: data = get_ebird_data("https://api.ebird.org/v2/data/obs/CA-ON/recent?back=30")
      ↪# Look back 30 days
      data
```

```
[ ]:      speciesCode      comName      sciName      locId \
0      rebwoo      Red-bellied Woodpecker      Melanerpes carolinus      L385611
1      merlin      Merlin      Falco columbarius      L7944575
2      whbnut      White-breasted Nuthatch      Sitta carolinensis      L26394771
3      bkcchi      Black-capped Chickadee      Poecile atricapillus      L26394771
4      eursta      European Starling      Sturnus vulgaris      L4140370
..      ...
250      eargre      Eared Grebe      Podiceps nigricollis      L2891330
251      limpki      Limpkin      Aramus guarauna      L27844578
252      chukar      Chukar      Alektoris chukar      L12821276
253      pinwar      Pine Warbler      Setophaga pinus      L131153
254      indbun      Indigo Bunting      Passerina cyanea      L131153

                                locName      obsDt \
0                                Wildwood Reservoir      2023-12-05 11:55
1      57-299 Ravenscrest Rd, Georgina CA-ON (44.2144,...      2023-12-05 11:54
2                                Heyden, Ontario      2023-12-05 11:52
3                                Heyden, Ontario      2023-12-05 11:52
4                                Queensville - Eves-Kinsley Residence      2023-12-05 11:50
..      ...
250                                Ipperwash Beach      2023-11-07 07:32
251                                Watson's Corners Road      2023-11-06 15:15
252                                Sufian St      2023-11-06 14:01
253      Point Pelee National Park (general location fo...      2023-11-06 06:53
254      Point Pelee National Park (general location fo...      2023-11-06 06:53

      howMany      lat      lng      obsValid      obsReviewed      locationPrivate \
0      1.0      43.242202      -81.044083      True      False      False
1      1.0      44.214377      -79.400671      True      False      True
2      1.0      46.642828      -84.306234      True      False      True
3      NaN      46.642828      -84.306234      True      False      True
4      8.0      44.136378      -79.449421      True      False      True
..      ...
250      1.0      43.206677      -81.995859      True      True      False
251      1.0      44.975124      -76.536940      True      True      True
252      1.0      45.819281      -77.220064      True      True      True
```

253	1.0	41.955399	-82.514000	True	True	False
254	1.0	41.955399	-82.514000	True	True	False

	subId	exoticCategory
0	S155793499	NaN
1	S155793460	NaN
2	S155793474	NaN
3	S155793474	NaN
4	S155793575	N
..
250	S153968504	NaN
251	S153926361	NaN
252	S154267581	X
253	S153897439	NaN
254	S153897439	NaN

[255 rows x 14 columns]

1.1.3 Attribute Analysis

Now that we have a solid set of results, we can proceed to analyze them towards the goal of understanding the range of species observed in Ontario in the past 30 days - and where they were observed.

First, we'll use the `pandas` programming package to examine key attributes of these observations. Then, we'll use `geopandas` - it's spatial data equivalent - to map them and assess their spatial distribution.

The following cell of code transforms our JSON-formatted data into a nicely formatted table.

```
[ ]: import pandas

data = pandas.DataFrame(data)

data
```

	speciesCode	comName	sciName	locId	\
0	rebwoo	Red-bellied Woodpecker	Melanerpes carolinus	L385611	
1	merlin	Merlin	Falco columbarius	L7944575	
2	whbnut	White-breasted Nuthatch	Sitta carolinensis	L26394771	
3	bkcchi	Black-capped Chickadee	Poecile atricapillus	L26394771	
4	eursta	European Starling	Sturnus vulgaris	L4140370	
..	
250	eargre	Eared Grebe	Podiceps nigricollis	L2891330	
251	limpki	Limpkin	Aramus guarauna	L27844578	
252	chukar	Chukar	Alectoris chukar	L12821276	
253	pinwar	Pine Warbler	Setophaga pinus	L131153	
254	indbun	Indigo Bunting	Passerina cyanea	L131153	

		locName	obsDt	\
0		Wildwood Reservoir	2023-12-05 11:55	
1	57-299 Ravenscrest Rd, Georgina CA-ON (44.2144,...	2023-12-05 11:54		
2		Heyden, Ontario	2023-12-05 11:52	
3		Heyden, Ontario	2023-12-05 11:52	
4	Queensville - Eves-Kinsley Residence	2023-12-05 11:50		
..		
250		Ipperwash Beach	2023-11-07 07:32	
251		Watson's Corners Road	2023-11-06 15:15	
252		Sufian St	2023-11-06 14:01	
253	Point Pelee National Park (general location fo...	2023-11-06 06:53		
254	Point Pelee National Park (general location fo...	2023-11-06 06:53		

	howMany	lat	lng	obsValid	obsReviewed	locationPrivate	\
0	1.0	43.242202	-81.044083	True	False	False	
1	1.0	44.214377	-79.400671	True	False	True	
2	1.0	46.642828	-84.306234	True	False	True	
3	NaN	46.642828	-84.306234	True	False	True	
4	8.0	44.136378	-79.449421	True	False	True	
..	
250	1.0	43.206677	-81.995859	True	True	False	
251	1.0	44.975124	-76.536940	True	True	True	
252	1.0	45.819281	-77.220064	True	True	True	
253	1.0	41.955399	-82.514000	True	True	False	
254	1.0	41.955399	-82.514000	True	True	False	

	subId	exoticCategory
0	S155793499	NaN
1	S155793460	NaN
2	S155793474	NaN
3	S155793474	NaN
4	S155793575	N
..
250	S153968504	NaN
251	S153926361	NaN
252	S154267581	X
253	S153897439	NaN
254	S153897439	NaN

[255 rows x 14 columns]

Next, we analyze the `sciName` column to figure out exactly how many different species were observed.

```
[ ]: data["sciName"].nunique()
```

[]: 255

So there were 255 different species identified over the past 30 days in southern Ontario (this number will vary depending on when this tutorial is run!)

What kinds of species were these? This will require us to make a different, separate call to eBird platform. This is not unusual - APIs typically provide very segmented information, requiring multiple calls to get the data you want. First, we will acquire more information for each species observed.

```
[ ]: observed = list(data["speciesCode"].unique()) # A list of each unique species
      ↪ observed and their eBird species code
observed_formatted = ""
for bird in observed:
    observed_formatted += bird+", "
observed_formatted = observed_formatted[:-1]
# In the above lines of code, we took our list of unique species observed and
      ↪ formatted it in the way the eBird API requires (a comma-separated list)

species_information = get_ebird_data("https://api.ebird.org/v2/ref/taxonomy/
      ↪ ebird?fmt=json&species="+observed_formatted)
species_information
```

```
[ ]:
      sciName      comName speciesCode \
0      Anser caerulescens      Snow Goose      snogoo
1      Anser rossii      Ross's Goose      rosgoo
2      Anser caerulescens x rossii      Snow x Ross's Goose (hybrid)      sxrgoo1
3      Anser anser      Graylag Goose      gragoo
4      Anser albifrons      Greater White-fronted Goose      gwfgoo
..      ...      ...      ...
250      Piranga rubra      Summer Tanager      sumtan
251      Cardinalis cardinalis      Northern Cardinal      norcar
252      Pheucticus ludovicianus      Rose-breasted Grosbeak      robgro
253      Passerina cyanea      Indigo Bunting      indbun
254      Spiza americana      Dickcissel      dickci
```

```
      category  taxonOrder  bandingCodes  comNameCodes  sciNameCodes  \
0      species      256      [SNGO]      []      [ANCA]
1      species      260      [ROGO]      []      [ANRO]
2      hybrid      261      [SRGH]      [SNGO, ROGO]      [ANRO, ANCA]
3      species      263      [GRGO]      []      [ANAN]
4      species      271      [GWFG]      []      [ANAL]
..      ...      ...      ...      ...      ...
250      species      33885      [SUTA]      []      [PIRU]
251      species      33967      [NOCA]      []      [CACA]
252      species      34009      [RBGR]      []      [PHLU]
253      species      34060      [INBU]      []      [PACY]
254      species      34080      [DICK]      []      [SPAM]
```

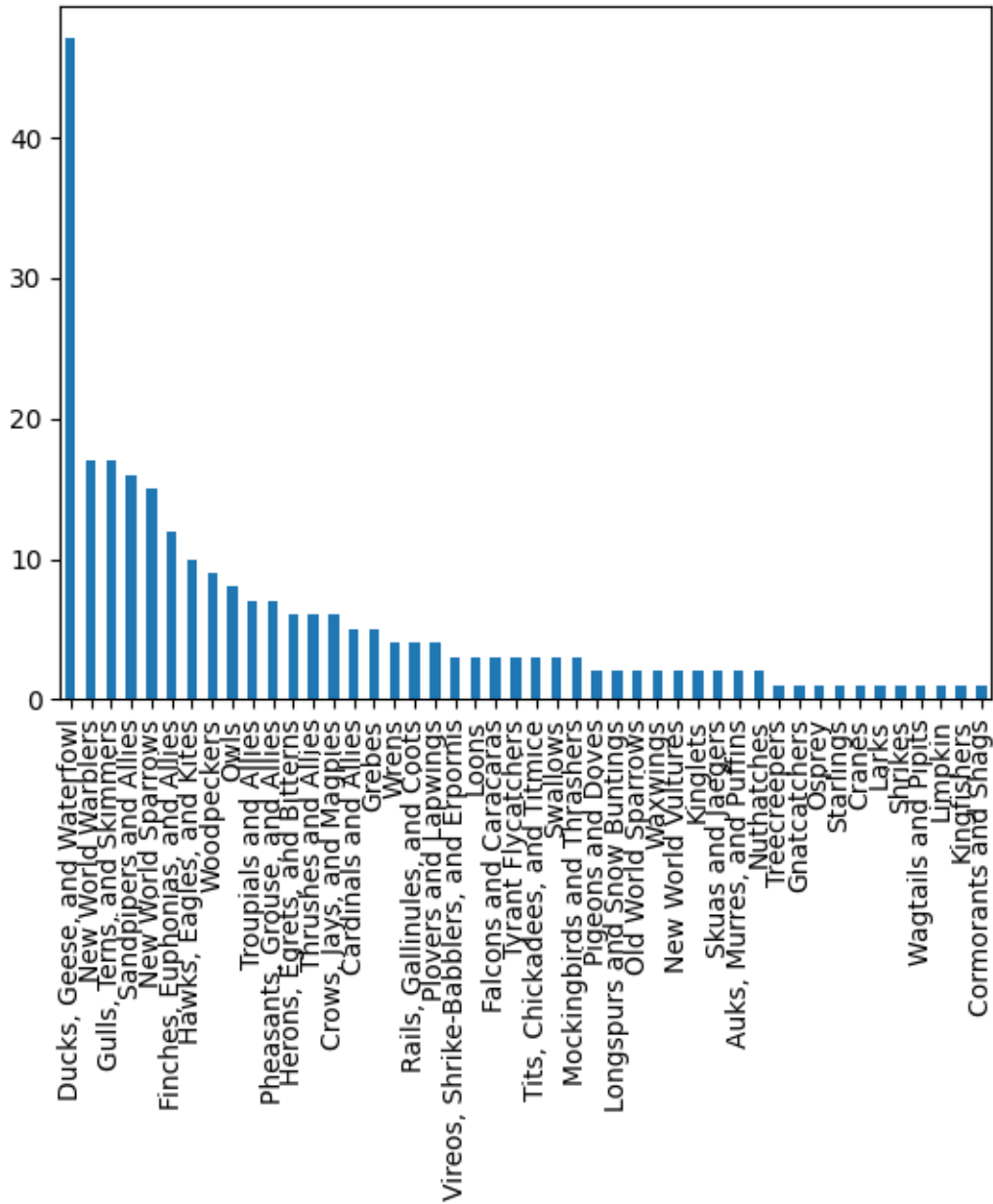

	order	familyCode	familyComName	familySciName
0	Anseriformes	anatid1	Ducks, Geese, and Waterfowl	Anatidae
1	Anseriformes	anatid1	Ducks, Geese, and Waterfowl	Anatidae
2	Anseriformes	anatid1	Ducks, Geese, and Waterfowl	Anatidae
3	Anseriformes	anatid1	Ducks, Geese, and Waterfowl	Anatidae
4	Anseriformes	anatid1	Ducks, Geese, and Waterfowl	Anatidae
..
250	Passeriformes	cardin1	Cardinals and Allies	Cardinalidae
251	Passeriformes	cardin1	Cardinals and Allies	Cardinalidae
252	Passeriformes	cardin1	Cardinals and Allies	Cardinalidae
253	Passeriformes	cardin1	Cardinals and Allies	Cardinalidae
254	Passeriformes	cardin1	Cardinals and Allies	Cardinalidae

[255 rows x 12 columns]

We can summarize this information and then plot it in a histogram.

```
[ ]: species_information['familyComName'].value_counts().plot(kind='bar')
```

```
[ ]: <Axes: >
```



To no one's surprise, we're mostly observing ducks, geese, and waterfowl!

Now let's see how representative this list of species is compared to ALL species ever observed in Ontario. We'll get that information from yet another part of eBird's API - that is, using another, different URL.

```
[ ]: species_list = get_ebird_data("https://api.ebird.org/v2/product/spplist/CA-ON")
species_list
```

```
[ ]:      0
0    bbwduc
1    fuwduc
2    bahgoo
3    empgoo
4    snogoo
..    ...
639  lazibun
640  indbun
641  varbun
642  paibun
643  dickci

[644 rows x 1 columns]
```

There have been many species observed over time in Ontario. In the past 30 days, we have observed the following percent of these species:

```
[ ]: str(round((species_information['speciesCode'].unique()/species_list[0].
↪shape[0]) * 100, 2)) + "%"
```

```
[ ]: '39.6%'
```

1.1.4 Spatial Analysis

Finally, let's look at WHERE these species were observed. Because of the nature of the eBird API, we don't have access to ALL observations in the recent past, so we can't really make claims about the spatial distribution of bird sightings (you would need to manually download the data from the Global Biodiversity Information Facility to do that). Nonetheless, we could consider what we do have - the most recent observation of each species observed in the past 30 days in Ontario - as perhaps representative of where citizen scientists are operating. However, it is crucial to remember that these data reflect where people who use eBird are finding birds. The data tell us as much about people as birds. We are likely to map eBird observations near cities, in parks, and so on.

First, we'll load the `geopandas` and `folium` programming packages then we'll load our data into the `geopandas` format and use `folium` to display them in an interactive map.

```
[ ]: import geopandas
import folium

map_data = geopandas.GeoDataFrame(data, geometry=geopandas.
↪points_from_xy(data['lng'], data['lat']), crs = 4326)

m = folium.Map()

birds_layer = folium.GeoJson(
    data = map_data,
```

```

        tooltip = folium.GeoJsonTooltip(fields=["comName", "sciName", "obsDt", "
↪"howMany"])
    ).add_to(m)

    bounds = m.get_bounds()
    m.fit_bounds(bounds, padding=0)

    m

```

```
[ ]: <folium.folium.Map at 0x7af56c5e93f0>
```

You may not be surprised to hear at this point that we can use another part of eBird’s API to get around this limitation to accessing bird observation data. While we may not be able to get all observations for all species all at once, we can use the “Recent observations of a species in a region” query to acquire a more robust sample of observations in an area for a single species, and by calling this query more than once, we can gather data for more than one species.

It is important to note that, per the eBird API documentation, we will still only be able to get the most recent observation made in each “location” in the region. What eBird considers a “location” is unclear. It uses various “codes” to describe locations. For instance, L227544 refers to a specific area near Cornell University. For our purposes, this just means that we will be sampling locations in the Ontario region with observations of a species or two in the past 30 days, but we will not be able to say anything specific about how many total observations were made.

So, let’s compare where two random species were observed. Keep in mind that there may not be that many observations of relatively rare species.

```

[ ]: # First, get two random species observed in Ontario in the past 30 days
    random = list(data.sample(2)["speciesCode"].unique())

    # Get more observations of these
    observations = pandas.DataFrame()
    for species in random:
        species_data = get_ebird_data("https://api.ebird.org/v2/data/obs/CA-ON/recent/
↪"+species+"?back=30")
        observations = pandas.concat([observations, species_data])

    # Map the observations
    ## Different colours will indicate different species, and different sizes of
    ↪circles will reflect how many individual birds were observed in that
    ↪location at that time
    colormap = {random[0]: "orange", random[1]: "blue"}
    def size(cnt):
        radius = 1
        if cnt < 10:
            radius = 8
        elif cnt < 20:
            radius = 12

```

```

elif cnt < 30:
    radius = 16
elif cnt < 50:
    radius = 20
else:
    radius = 32
return radius

m = folium.Map()

birds = [folium.CircleMarker(location=[bird["lat"], bird["lng"]],
    popup=folium.Popup(bird["comName"]+'<br><b>Scientific Name:</b>␣
↳'+bird["sciName"]+'<br><b>Location:</b> ' +bird["locName"]+'<br><b>Date most␣
↳recently observed:</b> ' +bird["obsDt"]),
    radius=size(bird["howMany"]), fill_color=colormap[bird["speciesCode"]],␣
↳color="black", weight=1, fill_opacity = 1) for index,bird in observations.
↳iterrows()]

for bird in birds:
    m.add_child(bird)

bounds = m.get_bounds()
m.fit_bounds(bounds, padding=0)

m

```

```
[ ]: <folium.folium.Map at 0x7af56eea7e80>
```

```
[ ]:
```